

EE5110/6110 Special Topics in Automation and Control

Autonomous Systems: Unmanned Aerial Vehicles

Modeling and Control System Design

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What is a **flight dynamics model**?

A flight dynamics model is a set of equations that explains

- 1. How aerodynamic forces are generated on the aircraft body?
- 2. How actuators and external factors affect these forces?
- 3. How these forces affect the aircraft's motion?





Why do we need a **flight dynamics model**?

1. Design and tuning of autonomous flight control law

Modern control techniques are mostly model based. An accurate flight dynamics model can assist flight control law design efficiently in terms of parameter tuning and performance evaluation.

2. Design and modification of aircraft structure

Aerodynamic stability, physical limits and characteristics can be analyzed based on the obtained model. Mechanical modification or re-structuring can be applied if there is inherent deficiencies in the original structure.



How do we obtain a **flight dynamics model**?

1) System Identification Approach

- Build mathematical model by fitting measurement data
- Simplified, linear, black-box
- Flight test experiments
- 2) First-principles Modeling Approach
 - Build mathematical model by manipulating equations of physics
 - Complex, nonlinear, rigorous with physical meanings
 - Test-bench experiments
- 3) Combination of the above two approaches

System Identification Approach Single-rotor helicopter







Despite the modeling approach adopted, we always need to first

- 1. Understand the basic working principles
- 2. Define coordinate frames
- *3. Define input and output variables* (*or state variables*)



System Identification Approach is suitable for obtaining linear dynamics model, and can be conducted in both time-domain and frequency-domain.



System Identification Approach Single-rotor helicopter



Toolkits:



IDENT







Step 1: Data collection and preprocessing:





Frequency sweep input signal



Step 1: Data collection and preprocessing:



System Identification Approach Single-rotor helicopter





System Identification Approach Single-rotor helicopter









Step 3: Unknown parameter identification:

- 1). Angular rate dynamics; 2). Horizontal velocity dynamics;
- 3). Yaw dynamics; 4). Heave dynamics;

a. Time-domain: Hover model of a single-rotor helicopter

		•••••					• • • • • • • • • • •	••••								
	-0.1778	0	0	0	0	-9.7807	-9.7807	0	0	0	0]	0	0	0	0]
	0	-0.3104	0	0	9.7807	0	0	9.7807	0	0	0		0	0	0	0
	-0.3326	-0.5353	0	0	0	0	75.764	343.86	0	0	0		0	0	0	0
	-0.1903	-0.2490	0	0	0	0	172.62	-59.958	0	0	0		0	0	0	0
	0	0	1	0	0	0	0	0	0	0	0		0	0	0	0
A =	0	0	0	1	0	0	0	0	0	0	0	, B=	0	0	0	0,
	0	0	0	-1	0	0	-8.1222	4.6535	0	0	0		0.0496	2.6224	0	0
	0	0	-1	0	0	0	-0.0921	-8.1222	0	0	0		2.4928	0.1741	0	0
	0	0	0	0	0	0	0	0	-0.6821	-0.0535	0		0	0	7.8246	0
	0	0	0	0	0	0	0	0	-0.2892	-5.5561	-36.674		0	0	1.6349	-58.4053
	0	0	0	0	0	0	0	0	0	2.7492	-11.112		0	0	0	0



b. Frequency-domain: Hovering model of a single-rotor helicopter

Extra metrics for parameter accuracy evaluation:

1). Cramer-Rao bound (CR% < 20%)	; 2). Insensitivity (I% < 10%)
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Parameter	Value	CR%	I%
X_u	-0.5198	20.31	9.827
X_{a_s}	-9.7807 !		
Y_v	-0.4225	19.51	9.231
Y_{bs}	9.7807 !		
L_u	-0.3262 !		
L_v	-0.0675 *		
L_{a_s}	0 +		
L_{b_s}	439.4000	2.489	1.020
M_u	-0.0814 *		
M_v	-2.1210	18.02	7.143
M_{a_s}	275.1000	2.794	1.067
M_{b_s}	-42.1100	14.17	4.562
τ	-4.7790	7.678	2.367
a_{b_s}	2.3850	13.67	4.309

b_{a_s}	0		
Z_w	-0.5996	19.33	12.13
Z_r	0.1103	6.369	2.514
N_r	-4.6150	7.515	2.441
N_{rfb}	154.5000 *		
N_v	0		
K_r	-1.0830	5.805	1.929
K_{rfb}	-9.2300 *		
A_{lat}	0.0610	16.62	8.979
A_{lon}	2.8080	2.731	1.319
B_{lat}	2.1810	2.687	1.202
B_{lon}	0 +		
Z_{col}	22.9400	3.776	1.856
N_{ped}	-154.5000	6.663	1.833

+ : Eliminated in model structure determination

! : Fixed empirically

* : Tied to other parameters



Frequency-domain: Hovering model of a single-rotor helicopter

Frequency responses matching



System Identification Approach Single-rotor helicopter



Step 4: Model validation





Rigid-body dynamics



G. Cai, B. M. Chen, and T. H. Lee, Unmanned Rotorcraft Systems, New York: Springer, 2011.



Main rotor forces and moments $T_{\rm mr}$ $X_{\rm mr} = -T\sin(a_{\rm s})$ $Y_{\rm mr}$ $Z_{\rm mr}$ $Y_{\rm mr} = T \sin(b_{\rm s})$ $M_{\rm mr}$ $Z_{\rm mr} = -T\cos(a_{\rm s})\cos(b_{\rm s})$ $L_{\rm mr} = (K_{\beta} + TH_{\rm mr})\sin(b_{\rm s})$ $X_{\rm mr}$ $M_{\rm mr} = (K_{\beta} + TH_{\rm mr})\sin(a_{\rm s})$ $L_{\rm mr}$ $N_{\rm mr} = -(P_{\rm pr} + P_{\rm i} + P_{\rm pa} + P_{\rm c})/\Omega$ $T_{\rm mr} = \frac{\rho \,\Omega_{\rm mr} \,R_{\rm mr}^2 \,C_{\rm l\alpha,mr} \,b_{\rm mr} \,c_{\rm mr}}{4} (w_{\rm bl,mr} - v_{\rm i,mr})$ $\geq N_{\rm mr}$ $v_{\rm i,mr}^2 = \sqrt{\left(\frac{\hat{v}_{\rm mr}^2}{2}\right)^2 + \left(\frac{T_{\rm mr}}{2\,\rho\,\pi\,R_{\rm mr}^2}\right)^2} - \frac{\hat{v}_{\rm mr}^2}{2}$ $\int P_{\rm pr} = \frac{\rho \Omega R^2 C_{\rm D0} b_{\rm mr} c_{\rm mr}}{8} \Big[(\Omega R)^2 + 4.6 (u_{\rm a}^2 + v_{\rm a}^2) \Big]$ $\hat{v}_{\rm mr}^2 = u_{\rm a}^2 + v_{\rm a}^2 + w_{\rm r,mr}(w_{\rm r,mr} - 2v_{\rm i,mr})$ $$\begin{split} P_{\rm i} &= T v_{\rm i} \\ P_{\rm c} &= -mg w_{\rm a} \\ P_{\rm pa} &= |X_{\rm fus} u_{\rm a}| + |Y_{\rm fus} v_{\rm a}| + |Z_{\rm fus} (w_{\rm a} - v_{\rm i})|. \end{split}$$ $w_{\rm r,mr} = w_{\rm a} + a_{\rm s} \, u_{\rm a} - b_{\rm s} \, v_{\rm a}$ $w_{\rm bl,mr} = w_{\rm r,mr} + \frac{2}{3}\Omega_{\rm mr} R_{\rm mr} \theta_{\rm col}$ Initial values: T(0) = mg, $v_i(0) = \frac{mg}{2\rho\pi B^2}$, $\hat{v}(0) = 0$. 17



Tail rotor forces and moments

$$T_{\rm tr} = \frac{\rho \Omega_{\rm tr} R_{\rm tr}^2 C_{\rm l\alpha, tr} b_{\rm tr} c_{\rm tr}}{4} (w_{\rm bl, tr} - v_{\rm i, tr})$$

$$v_{\rm i, tr}^2 = \sqrt{\left(\frac{\hat{v}_{\rm tr}^2}{2}\right)^2 + \left(\frac{T_{\rm tr}}{2\rho \pi R_{\rm tr}^2}\right)^2} - \frac{\hat{v}_{\rm tr}^2}{2}$$

$$\hat{v}_{\rm tr}^2 = (w_{\rm a} + qD_{\rm tr})^2 + u_{\rm a}^2 + w_{\rm r, tr} (w_{\rm r, tr} - 2v_{\rm i, tr})$$

$$w_{\rm r, tr} = v_{\rm a} - rD_{\rm tr} + pH_{\rm tr}$$

$$w_{\rm bl, tr} = w_{\rm r, tr} + \frac{2}{3}\Omega_{\rm tr} R_{\rm tr} \theta_{\rm ped}$$

$$\theta_{\rm ped} = K_{\rm ped}\bar{\delta}_{\rm ped} + \theta_{\rm ped, 0}$$
Initial values:
$$Y_{\rm tr}(0) = \frac{1}{\Omega D_{\rm tr}} \left(\frac{\rho \Omega^3 R^4 C_{\rm D0}}{8}\right)$$



nitial values:
$$Y_{\rm tr}(0) = \frac{1}{\Omega D_{\rm tr}} \left(\frac{\rho \Omega^3 R^4 C_{\rm D0} b_{\rm mr} c_{\rm mr}}{8} + \frac{m^2 g^2}{2\rho \pi R^2} \right), \quad v_{\rm i,tr}(0) = \frac{Y_{\rm tr}(0)}{2\rho \pi R_{\rm tr}^2}, \quad \hat{v}_{\rm tr}(0) = 0$$

 $Y_{\rm tr} = -T_{\rm tr}$ $L_{\rm tr} = -Y_{\rm tr}H_{\rm tr}$, $N_{\rm tr} = Y_{\rm tr}D_{\rm tr}$



Fuselage forces

$$\begin{aligned} X_{\rm fus} &= \begin{cases} -\frac{\rho}{2} S_{\rm fx} \, u_{\rm a} \, v_{\rm i,mr}, & \text{if } |u_{\rm a}| \leq v_{\rm i,mr} \\ -\frac{\rho}{2} S_{\rm fx} \, u_{\rm a} \, |u_{\rm a}|, & \text{if } |u_{\rm a}| > v_{\rm i,mr} \end{cases} \\ Y_{\rm fus} &= \begin{cases} -\frac{\rho}{2} S_{\rm fy} \, v_{\rm a} \, v_{\rm i,mr}, & \text{if } |v_{\rm a}| \leq v_{\rm i,mr} \\ -\frac{\rho}{2} S_{\rm fy} \, v_{\rm a} \, |v_{\rm a}|, & \text{if } |v_{\rm a}| > v_{\rm i,mr} \end{cases} \end{aligned}$$

$$Z_{\rm fus} = -\frac{\rho}{2} S_{\rm fz} \, \left(w_{\rm a} - v_{\rm i,mr} \right) \, \left| w_{\rm a} - v_{\rm i,mr} \right|$$





Vertical fin forces and moments

$$Y_{\rm vf} = \begin{cases} -\frac{\rho}{2} C_{\rm l\alpha,vf} S_{\rm vf} v_{\rm vf} |u_{\rm a}|, & \text{if } \left|\frac{v_{\rm vf}}{u_{\rm a}}\right| \le \tan(\alpha_{\rm st}) \\ -\frac{\rho}{2} S_{\rm vf} v_{\rm vf} |v_{\rm vf}|, & \text{if } \left|\frac{v_{\rm vf}}{u_{\rm a}}\right| > \tan(\alpha_{\rm st}) \text{ (stalled)} \end{cases}$$

where $v_{\rm vf} = v_{\rm a} + v_{\rm i,tr} - rD_{\rm vf}$

$$L_{\rm vf} = Y_{\rm vf} H_{\rm vf}, \quad N_{\rm vf} = Y_{\rm vf} D_{\rm vf}$$

Horizontal fin forces and moments

$$Z_{\rm hf} = \begin{cases} -\frac{\rho}{2} C_{\rm l\alpha, hf} S_{\rm hf} w_{\rm hf} |u_{\rm a}| , & \text{if } \left|\frac{w_{\rm hf}}{u_{\rm a}}\right| \le \tan(\alpha_{\rm st}) \\ -\frac{\rho}{2} S_{\rm hf} w_{\rm hf} |w_{\rm hf}| , & \text{if } \left|\frac{w_{\rm hf}}{u_{\rm a}}\right| > \tan(\alpha_{\rm st}) \text{ (stalled)} \end{cases}$$

where $w_{\rm hf} = w_{\rm a} + q D_{\rm hf} - v_{\rm i,mr}$







Direct measurement

Parameter	Physical meaning		
$I_{\rm mr}=0.055~{\rm kg}{\cdot}{\rm m}^2$	Main blade inertia w.r.t. rotor hub		
$I_{\rm sb}=0.004~{\rm kg}{\cdot}{\rm m}^2$	Stabilizer bar inertia w.r.t. rotor hub		
R = 0.7 m	Main rotor radius		
$R_{\rm sb,in}=0.231~{\rm m}$	Stabilizer bar inner radius		
$R_{\rm sb,out}=0.312~{\rm m}$	Stabilizer bar outer radius		
$R_{\rm tr}=0.128~{\rm m}$	Tail rotor radius		
$S_{\rm fx}=0.103~{\rm m}^2$	Effective longitudinal fuselage drag area		
$S_{\rm fy}=0.900~{\rm m}^2$	Effective lateral fuselage drag area		
$S_{\rm fz}=0.084~{\rm m}^2$	Effective vertical fuselage drag area		
$S_{\rm hf}=0.011~{\rm m}^2$	Horizontal fin area		
$S_{ m vf}=0.007~{ m m}^2$	Vertical fin area		
b=2	Main blade number		
$b_{\rm tr}=2$	Tail blade number		
$c=0.062~{\rm m}$	Main blade chord length		
$c_{\rm sb}=0.059~{\rm m}$	Stabilizer bar chord length		
$c_{\rm tr}=0.029~{\rm m}$	Tail rotor chord length		
$g=9.781\mathrm{N}{\cdot}\mathrm{kg}^{-1}$	Acceleration of gravity		
$m=9.750~{\rm kg}$	Helicopter mass		
$n_{\rm tr} = 4.650$	Gear ratio of the tail rotor to the main rotor		
$\rho=1.290~\rm kg/m^3$	Air density		

Ground Tests

Parameter	Physical meaning	
$D_{12} = 0.751 \text{ m}$	Horizontal fin location behind the CG	
$D_{\rm hf} = 0.751 {\rm m}$	Tail solor bub location behind the CG	CC
$D_{\rm tr} = 0.084 {\rm m}$	Vertical fin location behind the CG	LG
$H_{\rm vf} = 0.337 \rm{m}$	Main rotor hub location above the CG	Location
$H_{\rm mr} = 0.172 {\rm m}$	Tail rotor hub location above the CG	
$H_{\rm vf} = 0.112 {\rm m}$ $H_{\rm vf} = 0.184 {\rm m}$	Vertical fin location above the CG	
$A_{\rm lon}=0.210$ rad	Direct linkage gain from δ_{lon} to main blade deflection	on
$B_{ m lat}=0.200$ rad	Direct linkage gain form δ_{lat} to main blade deflection	m Airfoil
$C_{ m lon}=0.560$ rad	Linkage gain from δ_{lon} to stabilizer bar deflection	deflection
$D_{ m lat}=0.570~ m rad$	Linkage gain from δ_{lat} to stabilizer bar deflection	uejiectioi
$K_{\rm sb} = 1$	Ratio of main blade deflection to stabilizer bar TPP	titling angle
$I_{\rm xx}=0.249~{\rm kg}{\cdot}{\rm m}^2$	Rolling moment of inertia	
$I_{\rm yy}=0.548~{\rm kg}{\cdot}{\rm m}^2$	Pitching moment of inertia	Inertia
$I_{\rm zz}=0.787~{\rm kg}{\cdot}{\rm m}^2$	Yawing moment of inertia	moreta
$K_{\mathrm{I}}K_{\mathrm{a}}=8.499$ rad	Product of integral gain $K_{\rm I}$ and scaling factor $K_{\rm a}$	
$K_{ m P} K_{ m a} = 1.608$ rad	Product of proportional gain $K_{\rm P}$ and scaling factor	Ka
$K_{ m col}=-0.165$ rad	Ratio of $ heta_{ m col}$ to $\delta_{ m col}$	Collective
$K_{\rm ped} = 1$	Ratio of $ heta_{ m ped}$ to $ar{\delta}_{ m ped}$	Pitch
$ heta_{ m col,0}=0.075$ rad	$ heta_{ m col}$ value when $\delta_{ m col}$ is zero	FILLI
$\theta_{\rm rad} = 0.143$ rad	θ_{-1} value when $\overline{\delta}_{-1}$ is zero	curve



Wind Tunnel	Data	Actual Flight Tests			
Parameter	Physical meaning	Parameters	Physical meaning		
$C_{l\alpha,hf} = 2.85 \text{ rad}^{-1}$ $\hat{C}_{l\alpha,mr} = 5.71 \text{ rad}^{-1}$ $\hat{C}_{l\alpha,sb} = 2.23 \text{ rad}^{-1}$ $\hat{C}_{l\alpha,tr} = 2.23 \text{ rad}^{-1}$ $C_{l\alpha,vf} = 2.85 \text{ rad}^{-1}$	Horizontal fin lift curve slope Main blade lift curve slope Stabilizer bar lift curve slope Tail blade lift curve slope Vertical fin lift curve slope	$\begin{array}{l} A_{\rm b_{s}} = 9.720 \; {\rm sec^{-1}} \\ B_{\rm a_{s}} = 10.704 \; {\rm sec^{-1}} \\ K_{\rm I} = 2.2076 \\ K_{\rm P} = 0.4177 \\ K_{\rm a} = -3.85 \; {\rm rad} \\ K_{\beta} = 112.84 \; {\rm N} \cdot {\rm m} \\ C_{\rm l\alpha,mr} = 5.73 \; {\rm rad^{-1}} \\ C_{\rm l\alpha,sb} = 2.13 \; {\rm rad^{-1}} \\ C_{\rm l\alpha,tr} = 2.81 \; {\rm rad^{-1}} \end{array}$	Coupling effect Coupling effect Integral gain of the embedded PI controller Proportional gain of the embedded PI controller Scaling factor of the amplifier circuit Main rotor spring constant Main blade lift curve slope Stabilizer bar lift curve slope Tail blade lift curve slope		

 $\Omega_{\rm mr} = 193.73~{\rm rad}$

 $\Omega_{tr} = 900.85 \text{ rad}$

Main rotor rotating speed

Main rotor rotating speed

First-principles Modeling Approach Coaxial & quadrotor helicopters





First-principles Modeling Approach Quadrotor working principles





- Symmetric structure
- All types of motion controlled by adjusting motor speeds
- All thrust forces in the UAV body frame z-axis direction
- Open-loop dynamics fast and unstable

First-principles Modeling Approach Quadrotor force & torque decomposition





First-principles Modeling Approach Coaxial helicopter working principles



- Dual main rotors spin in opposite directions
- Rotational speed can be controlled
 - Sum \rightarrow Heave motion
 - Difference \rightarrow Yaw motion
- Bottom rotor attached to swashplate
 - Servo 1 \rightarrow Roll motion
 - Servo 2 \rightarrow Pitch motion
- > Top rotor attached to stabilizer bar
 - Roll & Pitch motion damped



First-principles Modeling Approach Coaxial helicopter working principles (stabilizer bar)





- Roll and pitch dynamics slowed down
- Stability increased (inherently stable)
- Maneuverability decreased

First-principles Modeling Approach Coaxial helicopter force & torque decomposition









First-principles Modeling Approach Coaxial helicopter model structure





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First-principles Modeling Approach Kinematics





First-principles Modeling Approach Rigid-body dynamics





First-principles Modeling Approach Mass & moment of inertia





m = 0.934 kg

 $J_{zz} = \frac{mgl_1l_2l_3t_1^2}{4\pi^2 l} \cdot \frac{l_1\sin\alpha_1 + l_2\sin\alpha_2 + l_3\sin\alpha_3}{l_2l_3\sin\alpha_1 + l_1l_3\sin\alpha_2 + l_1l_2\sin\alpha_3}$

First-principles Modeling Approach Rotor thrust and torque generation







First-principles Modeling Approach Motor dynamics







$$\dot{\Omega}_{up} = \frac{1}{\tau_{mt}} (m_{up} \delta_{up} + \Omega_{up}^* - \Omega_{up})$$
$$\dot{\Omega}_{dw} = \frac{1}{\tau_{mt}} (m_{dw} \delta_{dw} + \Omega_{dw}^* - \Omega_{dw})$$

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$$\overline{\delta}_{rud} = K_{P}(K_{a}\delta_{rud} - r) + K_{I}r_{fb} \qquad \qquad \delta_{up} = \delta_{thr} + \overline{\delta}_{rud}$$
$$\dot{r}_{fb} = K_{a}\delta_{rud} - r \qquad \qquad \delta_{dw} = \delta_{thr} - \overline{\delta}_{rud}$$

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First-principles Modeling Approach Tip-path-plane dynamics





 $a_{dw} = A_{a,dw} \delta_{ele} + A_{b,dw} \delta_{ail}$





 $\dot{a}_{up} = -\frac{1}{\tau_{up}}a_{up} - A_{a,up}q - A_{b,up}p$



 $b_{\rm dw} = \underline{B}_{b,\rm dw} \,\delta_{\rm ail} + B_{a,\rm dw} \,\delta_{\rm ele}$



 $\dot{b}_{up} = -\frac{1}{\tau_{up}} b_{up} - \frac{B_{b,up}}{B_{b,up}} p - B_{a,up} q \qquad 37$

First-principles Modeling Approach Tip-path-plane dynamics



$$\begin{pmatrix} p \\ q \\ q \\ a_{up} \\ b_{up} \end{pmatrix} = \begin{bmatrix} \frac{-X_{dw}B_{p,dw}}{J_{xx}} & 0 & 0 & \frac{X_{up}}{J_{xx}} \\ 0 & \frac{-X_{dw}A_{q,dw}}{J_{yy}} & \frac{X_{up}}{J_{yy}} & 0 \\ -A_{b,up} & -A_{a,up} & -\frac{1}{\tau_{sb}} & 0 \\ -B_{b,up} & -B_{a,up} & 0 & -\frac{1}{\tau_{sb}} \end{bmatrix} \begin{pmatrix} p \\ q \\ a_{up} \\ b_{up} \end{pmatrix} + \begin{bmatrix} \frac{X_{dw}B_{b,dw}}{J_{xx}} & \frac{X_{dw}A_{a,dw}}{J_{xy}} \\ \frac{X_{dw}A_{b,dw}}{J_{yy}} & \frac{X_{dw}A_{a,dw}}{J_{yy}} \\ 0 & 0 \\ 0 & 0 \end{bmatrix} \begin{pmatrix} \delta_{ail} \\ \delta_{ele} \end{pmatrix}$$

$$\begin{bmatrix} p \\ q \\ a_{up} \\ b_{up} \end{bmatrix} = \begin{bmatrix} -17.19 & 0 & 0 & 934.1 \\ 0 & -5.360 & 291.3 & 0 \\ 0.2745 & -0.49 & -5 & 0 \\ -0.49 & -0.2745 & 0 & -5 \end{bmatrix} \begin{bmatrix} p \\ q \\ a_{up} \\ b_{up} \end{bmatrix} + \begin{bmatrix} -102.48 & -38.08 \\ -11.73 & 31.95 \\ 0 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} \delta_{ail} \\ \delta_{ele} \end{bmatrix}$$









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Parameter identification via elevator frequency sweep





- 1. Direct measurement
- 2. Test-bench experiment
- 3. Flight test data

$\rho = 1.204$	$m_{\rm up} = 106.90$	$\Omega_{up}^* = 203.38$
m = 0.977	$m_{\rm dw} = 106.45$	$\Omega_{\rm dw}^{*} = 217.88$
R = 0.250	$\tau_{\rm sb} = 0.2$	$K_{\beta}^{am} = 4.377$
g = 9.781	$\tau_{\rm mt} = 0.12$	'
$S_{fx} = 0.00835$	$K_{a} = 6.4267$	$J_{xx} = 0.0059$
$S_{fy} = 0.01310$	$K_{\mathrm{P}} = 0.667/K_{\mathrm{a}}$	$J_{yy} = 0.0187$
$S_{fz} = 0.01700$	$K_{\rm I} = 0.713/K_{\rm a}$	$J_{zz} = 0.0030$
$l_{up} = 0.195$	$J_{\rm up} = 6.8613 \cdot 10^{-4}$	$A_q = 0.0204$
$l_{dw} = 0.120$	$J_{\rm dw} = 3.2906 \cdot 10^{-4}$	$B_p = 0.0204$
$k_{T,up} = 1.23 \cdot 10^{-4}$	$A_{a,up} = 0.4900$	$A_{a,\mathrm{dw}} = 0.1217$
$k_{T,dw} = 8.50 \cdot 10^{-5}$	$A_{b,up} = -0.2745$	$A_{b,dw} = -0.0450$
$k_{Q,up} = 4.23 \cdot 10^{-6}$	$B_{a,up} = 0.2745$	$B_{a,dw} = -0.0450$
$k_{Q,dw} = 3.68 \cdot 10^{-6}$	$B_{b,up} = 0.4900$	$B_{b,dw} = -0.1217$

UAV Control System Design What is a control system?





Objective: To make the system **OUTPUT** and the desired **REFERENCE** as close as possible, i.e., to make the **ERROR** as small as possible.

Key Issues:(1) How to describe the system to be controlled? (Modeling)(2) How to design the controller? (Control)



There are many other factors of life have to be carefully considered when

designing a control system for real-life problems. These factors include:



These real-life factors, i.e., disturbances, uncertainties, nonlinearities, give birth of many modern control techniques...



The following is my personal view on the clarification of control techniques:

- Classical control: PID control, developed in 1940s and utilized heavily for in industrial processes. Examples: everywhere in life...
- Optimal control: Linear quadratic regulator (LQR), H₂ optimal control, Kalman filter, developed in 1960s to achieve certain optimal performance and boomed by NASA Apollo Project.
- Robust control: H_{∞} control, developed in 1980s and 90s to handle systems with uncertainties and disturbances and with high performances.
- Nonlinear control: Still on-going research topics, developed to handle nonlinear systems with high performances.
- Intelligent control: Knowledge-based control, adaptive control, neural and fuzzy control, etc., researched heavily in 1990s, developed to handle systems with unknown models. Examples: economic systems, social systems...





- Inner loop control is to guarantee the stability of the aircraft attitude
- Outer loop control is to control the aircraft position
- Flight schedule is to generate references for flight missions



Detailed structure of inner- and outer-loop control





Inner-loop control is to stabilize the overall aircraft and to control its attitude

We adopt the design specifications set by the USA military organization for military rotorcraft, which place strict requirements on:

- closed-loop stability
- bandwidth of pitch and roll attitude response
- coupling effect between roll and pitch channels
- coupling effect from heave control to yaw response
- disturbance rejection
- quickness of pitch, roll and yaw responses
- attitude hold for spike disturbance input



We aim to achieve **Level 1 performance** in all specifications in accordance with the military standards set by U.S. Army Aviation (ADS-33E-PRF).



For the inner loop, we treat wind gusts as disturbance entering the system and then formulate the overall problem as an H_{∞} control problem, which is to design a control law such that when it is applied to the given system, the resulting closed-loop system is stable and the effect of the disturbance to the output to be controlled (the position of the aircraft in our case) is minimized in H_{∞} sense:





 H_{∞} optimization is closely and commonly related to *robust control*...

Introduction to H_{∞} *control*



Consider a stabilizable and detectable linear time-invariant system Σ with a proper controller Σ_c



where

$$\Sigma: \begin{cases} \dot{x} = A \ x + B \ u + E \ w \\ y = C_1 \ x + & + D_1 \ w \\ z = C_2 \ x + & + D_2 \ u \end{cases}$$

- state variable
- y
- ⇔ measurement
 ⇔ controlled output

$$\Sigma_c: \begin{cases} \dot{v} = A_{\rm cmp} \ v + B_{\rm cmp} \ y \\ u = C_{\rm cmp} \ v + D_{\rm cmp} \ y \end{cases}$$

- control input \Leftrightarrow U
- \Leftrightarrow disturbance W
- controller state \Leftrightarrow V



The problem of H_{∞} control is to design a control law Σ_{c} such that when it is applied to the given plant with disturbance, i.e., Σ , we have

- The resulting closed loop system is internally stable (this is necessary for any control system design).
- The H_{∞} -norm of the resulting closed-loop transfer function from the disturbance to the controlled output is as small as possible, i.e., the effect of the disturbance on the controlled output is minimized.

Note: A transfer function is a function of frequencies ranging from 0 to ∞ . It is hard to tell if it is large or small. The common practice is to measure its norms instead. H_2 -norm and H_{∞} -norm are two commonly used norms in measuring the sizes of a transfer function.

UAV Control System Design Interpretation of H_{∞} norm



Given a stable and proper transfer function $T_{zw}(s)$, its H_{∞} -norm is defined as

$$\|T_{zw}\|_{\infty} = \sup_{0 \le \omega < \infty} \sigma_{\max} \left[T_{zw}(j\omega) \right]$$

where $\sigma_{\max}[T_{zw}(j\omega)]$ denotes the maximum singular value of $T_{zw}(j\omega)$. For a SISO transfer function $T_{zw}(s)$, it is equivalent to the magnitude of $T_{zw}(j\omega)$. Graphically,



Note: The H_{∞} -norm is the worst case gain in $T_{zw}(s)$. Thus, minimization of the H_{∞} -norm of $T_{zw}(s)$ is equivalent to the minimization of the worst case (gain) situation on the effect from the disturbance w to the controlled output z.



Most results in H_{∞} control deal with a so-called a regular problem or regular case because it is simple. An H_{∞} control problem is said to be **regular** if the following conditions are satisfied,

- 1. D_2 is of maximal column rank, i.e., D_2 is a tall and full rank matrix
- 2. The subsystem (A,B, C_2 , D_2) has no invariant zeros on the imaginary axis;
- 3. D_1 is of maximal row rank, i.e., D_1 is a fat and full rank matrix
- 4. The subsystem (A, E, C_1, D_1) has no invariant zeros on the imaginary axis.

An H_{∞} control problem is said to be **singular** if it is not regular, i.e., at least one of the above 4 conditions is not satisfied.

Solution to regular H_{∞} state feedback problem



Given $\gamma > \gamma_{\infty}^*$ (see the note below), solve the following algebraic Riccati equation

 $A^{\mathrm{T}}P + PA + C_{2}^{\mathrm{T}}C_{2} + PEE^{\mathrm{T}}P / \gamma^{2} - (PB + C_{2}^{\mathrm{T}}D_{2}) (D_{2}^{\mathrm{T}}D_{2})^{-1} (D_{2}^{\mathrm{T}}C_{2} + B^{\mathrm{T}}P) = 0$

for a unique positive semi-definite solution $P \ge 0$. The H_{∞} state feedback law is then given by

$$u = F \ x = -(D_2^{\mathrm{T}} D_2)^{-1} (D_2^{\mathrm{T}} C_2 + B^{\mathrm{T}} P) x$$

The resulting closed-loop system $T_{zw}(s)$ has the following property: $\|T_{zw}\|_{\infty} < \gamma$.

Note: The computation of the best achievable H_{∞} attenuation level, γ_{∞}^* , is very complicated. For certain cases, γ_{∞}^* can be computed exactly. Generally, γ_{∞}^* can only be obtained using some iterative algorithms. One method is to keep solving the Riccati equation for different values of γ until it hits γ_{∞}^* for which and any $\gamma < \gamma_{\infty}^*$, the Riccati equation does not have a solution. See Chen (2000) for details.



Solution to singular H_{∞} state feedback problem

Step 1: Given a $\gamma > \gamma_{\infty}^*$, choose $\varepsilon = 1$.

Step 2: Define the corresponding \tilde{C}_{γ} and \tilde{D}_{γ}

Step 3: Solve the following Riccati equation for \tilde{P} :

 $A^{\mathrm{T}}\tilde{P} + \tilde{P}A + \tilde{C}_{2}^{\mathrm{T}}\tilde{C}_{2} + \tilde{P}EE^{\mathrm{T}}\tilde{P} / \gamma^{2} - (\tilde{P}B + \tilde{C}_{2}^{\mathrm{T}}\tilde{D}_{2}) (\tilde{D}_{2}^{\mathrm{T}}\tilde{D}_{2})^{-1} (\tilde{D}_{2}^{\mathrm{T}}\tilde{C}_{2} + B^{\mathrm{T}}\tilde{P}) = 0$

Step 4: If $\tilde{P} > 0$, go to Step 5. Otherwise, reduce the value of ε and go to Step 2.

Step 5: Compute the required state feedback control law

 $u = F x = -(D_2^{'} D_2^{'})^{-1} (D_2^{'} C_2 + B^{'} P) x$

The resulting closed-loop system $T_{zw}(s)$ has: $\|T_{zw}\|_{\infty} < \gamma$.

More general results for the singular case can be found in Chen (2000).



$$\tilde{C}_2 := \begin{bmatrix} C_2 \\ \varepsilon I \\ 0 \end{bmatrix} \text{ and } \tilde{D}_2 := \begin{bmatrix} D_2 \\ 0 \\ \varepsilon I \end{bmatrix}$$

$$\mu = \tilde{F} \quad r = -(\tilde{D}^{\mathrm{T}} \tilde{D})^{-1} (\tilde{D}^{\mathrm{T}} \tilde{C} + R^{\mathrm{T}} \tilde{P}) r$$

Inner-loop control system design setup

$\dot{\mathbf{x}} = A\mathbf{x} + B\mathbf{u} + E\mathbf{w}$												
$-\mathbf{y} = C_1 \mathbf{x}$		۲O	0	1	0	0		0	0.00	09	0	0٦
		0	0	0	0.9992	0		0	-0.03	89	0	0
$\mathbf{h}_{\text{out}} = C_{\text{out}} \mathbf{x}$		0	0	-0.0302	-0.0056	-0.0003	585	.1165	11.44	48 –	59.529	0
out		0	0	0	-0.0707	267.7499	-0.	.0003		0	0	0
	A =	0	0	0	-1.0000	-3.3607	2.	.2223		0	0	0
$\langle \rangle$		0	0	-1	0	2.4483	-3	.3607		0	0	0
$\left(\begin{array}{c} \varphi \end{array} \right)$		0	0	0.0579	0.0108	0.0049	0.	.0037	-21.95	57	114.2	0
$ \rho $		0	0	0	0	0		0	-	-1	0	0
		0	0	0	0.0389	0		0	0.99	92	0	0
\mathbf{p}									0	0		0
$\mathbf{y} = \begin{bmatrix} a \end{bmatrix}$		Γ	0	()	07		Γ	0	0		0]
			0	()	0			0	0		0
r			0	() 43.36	35		-0.0	001 (.1756	-0.03	95
			0	()	0		0.0	000 0	.0003	0.03	38
(ψ)	B =	0.	2026	2.5878	8	0	E =		0	0		0
		2.	5878	-0.0663	3	0			0	0		0
(ϕ)			0	() -83.18	83		-0.0	002 -0	.3396	0.64	24
Υ			0	() -3.85	00			0	0		0
$\mathbf{h}_{\text{out}} = \theta $		L	0	()	0		L	0	0		0
(ψ)												

 H_{∞} state feedback control law...

$$\mathbf{u} = F \,\mathbf{x} + G \left(\mathbf{r} - \mathbf{h}_{\text{out,trim}}\right)$$

$$F = \begin{bmatrix} -1.0368 & -0.0604 & -0.0230 & -0.0083 & -0.2857 & -2.6165 & -0.0312 & 0.0499 & -0.0746 \\ 0.0760 & -0.9970 & 0.0174 & -0.0378 & -1.8340 & -0.1130 & 0.0026 & 0.0024 & -0.0169 \\ -0.0002 & -0.0185 & -0.0066 & 0.0004 & 0.0353 & 0.0990 & 0.0044 & 0.2295 & 0.2441 \end{bmatrix}$$

$$G = \left[C_{\text{out}} \left(A - BF\right)^{-1}B\right]^{-1} = \begin{bmatrix} 1.0368 & 0.0604 & 0.0746 \\ -0.0760 & 0.9970 & 0.0169 \\ 0.0002 & 0.0185 & -0.2441 \end{bmatrix}$$

The above state feedback control law is to be implemented together with a properly designed reduced-order observer for the state variables that cannot be measured...

Inner-loop control performance evaluation

Evaluation results with the standard set by U.S. Army Aviation...

UAV Control System Design Inner-loop command generator

The outer-loop control is to control the position of the aircraft and at the same

time to generate necessary commands for the inner-loop control system...

UAV Control System Design Properties of the virtual actuator

It can also be verified that coupling among each channel of the outer loop dynamics is very weak and thus can be ignored. As a result, all the x, y and z channels of the rotorcraft dynamics can be treated as decoupled and each channel can be characterized by

$$\begin{pmatrix} \dot{p}_* \\ \dot{v}_* \end{pmatrix} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \begin{pmatrix} p_* \\ v_* \end{pmatrix} + \begin{pmatrix} 0 \\ 1 \end{pmatrix} a_*$$

where p_* is the position, v_* is the velocity and a_* is the acceleration, which is treated a control input in our formulation.

For such a simple system, it can be controlled by almost all the control techniques available in the literature, which include the most popular and the simplest one such as PID control...

Control technique adopted for the outer loop

The outer-loop control system is designed using the so-called **robust** and perfect tracking (RPT) control technique developed by Chen and his co-workers. It is to design a controller such that the resulting closed-loop system is stable and the controlled output almost perfectly tracks a given reference signal in the presence of any initial conditions and external disturbances.

One of the most interesting features in the RPT control method is its capability of utilizing all possible information available in its controller structure. Such a feature is highly desirable for flight missions involving complicated maneuvers, in which not only the position reference is useful, but also its velocity and even acceleration information are important or even necessary to be used in order to achieve a good overall performance.

The RPT control renders flight formation of multiple UAVs a trivial task.

Introduction to RPT control

Consider the following continuous-time system:

 $\Sigma : \begin{cases} \dot{x} = A \ x + B \ u + E \\ y = C_1 \ x & + D_1 \\ h = C_2 \ x + D_2 \ u + D_{22} \end{cases} \begin{pmatrix} w, & x(0) = x_0, \\ w, & w \\ w \end{pmatrix}$ (8.1)

where $\mathbf{x} \in \mathbb{R}^n$ is the state, $\mathbf{u} \in \mathbb{R}^m$ is the control input, $\mathbf{w} \in \mathbb{R}^q$ is the external disturbance, $\mathbf{y} \in \mathbb{R}^p$ is the measurement output, and $\mathbf{h} \in \mathbb{R}^\ell$ is the output to be controlled. Given the external disturbance $\mathbf{w} \in L_p$, $\mathbf{p} \in [1, \infty)$, and any reference signal vector $\mathbf{r} \in \mathbb{R}^\ell$ with $\mathbf{r}, \dot{\mathbf{r}}, \dots, \mathbf{r}^{(\kappa-1)}, \kappa \ge 1$, being available, and $\mathbf{r}^{(\kappa)}$ being either a vector of delta functions or in L_p , the RPT problem for the system in (8.1) is to find a parameterized dynamic measurement control law of the following form:

$$\begin{cases} \dot{\boldsymbol{v}} = A_{\rm cmp}(\boldsymbol{\varepsilon})\boldsymbol{v} + B_{\rm cmp}(\boldsymbol{\varepsilon})\boldsymbol{y} + G_0(\boldsymbol{\varepsilon})\boldsymbol{r} + \dots + G_{\kappa-1}(\boldsymbol{\varepsilon})\boldsymbol{r}^{(\kappa-1)} \\ \boldsymbol{u} = C_{\rm cmp}(\boldsymbol{\varepsilon})\boldsymbol{v} + D_{\rm cmp}(\boldsymbol{\varepsilon})\boldsymbol{y} + H_0(\boldsymbol{\varepsilon})\boldsymbol{r} + \dots + H_{\kappa-1}(\boldsymbol{\varepsilon})\boldsymbol{r}^{(\kappa-1)} \end{cases}$$
(8.2)

such that when the controller of (8.2) is applied to the system of (8.1), we have the following

- 1. There exists an $\varepsilon^* > 0$ such that the resulting closed-loop system with $\mathbf{r} = 0$ and $\mathbf{w} = 0$ is asymptotically stable for all $\varepsilon \in (0, \varepsilon^*]$.
- 2. Let $\boldsymbol{h}(t, \boldsymbol{\varepsilon})$ be the closed-loop controlled output response and let $\boldsymbol{e}(t, \boldsymbol{\varepsilon})$ be the resulting tracking error, i.e., $\boldsymbol{e}(t, \boldsymbol{\varepsilon}) := \boldsymbol{h}(t, \boldsymbol{\varepsilon}) \boldsymbol{r}(t)$. Then, for any initial condition of the state, $\boldsymbol{x}_0 \in \mathbb{R}^n$,

$$\|\boldsymbol{e}\|_{\boldsymbol{p}} = \left(\int_0^\infty |\boldsymbol{e}(t)|^{\boldsymbol{p}} \, \mathrm{d}t\right)^{1/\boldsymbol{p}} \to 0 \text{ as } \boldsymbol{\varepsilon} \to 0.$$
(8.3)

RPT Control

Robust to

disturbance

66

The necessary and sufficient condition for the solvability of the general RTP control problem can be found in Chen (200). For the case when $D_1 = 0$, the conditions are relatively simple and are given as

(A, B) is stabilizable and (A, C₁) is detectable.
 D₂₂ = 0.
 (A, B, C₂, D₂) is right invertible and of minimum phase.
 Ker (C₂) ⊃ Ker (C₁).

where Ker(*X*) represents the null space of a constant matrix *X*.

We note that the last condition is automatically satisfied if the control output **h** of the given system is part of its measurement output **y**.

V

Solution to state feedback RPT control problem

Step 1. For a sufficiently small scalar ε_0 , we define

$$\tilde{\boldsymbol{C}}_{2} = \begin{bmatrix} \boldsymbol{C}_{2} \\ \boldsymbol{\varepsilon} I_{\kappa\ell+n} \\ 0 \end{bmatrix}, \quad \tilde{\boldsymbol{D}}_{2} = \begin{bmatrix} \boldsymbol{D}_{2} \\ 0 \\ \boldsymbol{\varepsilon} I_{m} \end{bmatrix},$$
$$\tilde{\boldsymbol{A}}_{0} = -\boldsymbol{\varepsilon}_{0} I_{\kappa\ell} + \begin{bmatrix} 0 & I_{\ell} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & I_{\ell} \\ 0 & 0 & \cdots & 0 \end{bmatrix}$$

Step 2. Then, solve for positive-definite solution for the following Riccati equation

$$\boldsymbol{P}\tilde{\boldsymbol{A}} + \tilde{\boldsymbol{A}}^{\mathrm{T}}\boldsymbol{P} + \tilde{\boldsymbol{C}}_{2}^{\mathrm{T}}\tilde{\boldsymbol{C}}_{2} - \left(\boldsymbol{P}\boldsymbol{B} + \tilde{\boldsymbol{C}}_{2}^{\mathrm{T}}\tilde{\boldsymbol{D}}_{2}\right) \left(\tilde{\boldsymbol{D}}_{2}^{\mathrm{T}}\tilde{\boldsymbol{D}}_{2}\right)^{-1} \left(\boldsymbol{P}\boldsymbol{B} + \tilde{\boldsymbol{C}}_{2}^{\mathrm{T}}\tilde{\boldsymbol{D}}_{2}\right)^{\mathrm{T}} = 0$$

Step 3. The required state feedback control law that solves the RPT problem is

$$\boldsymbol{u} = F(\boldsymbol{\varepsilon})\boldsymbol{x} + H_0(\boldsymbol{\varepsilon})\boldsymbol{r} + \dots + H_{\kappa-1}(\boldsymbol{\varepsilon})\boldsymbol{r}^{(\kappa-1)}$$

where $\tilde{\boldsymbol{F}}(\boldsymbol{\varepsilon}) = -(\tilde{\boldsymbol{D}}_2^{\mathrm{T}}\tilde{\boldsymbol{D}}_2)^{-1}(\boldsymbol{P}\boldsymbol{B} + \tilde{\boldsymbol{C}}_2^{\mathrm{T}}\tilde{\boldsymbol{D}}_2)^{\mathrm{T}} = [H_0(\boldsymbol{\varepsilon}) \quad \dots \quad H_{\kappa-1}(\boldsymbol{\varepsilon}) \quad F(\boldsymbol{\varepsilon})].$

$$a_{x,n} = -\left[\frac{\omega_{n,x}^{2}}{\varepsilon_{x}^{2}} \quad \frac{2\zeta_{x}\omega_{n,x}}{\varepsilon_{x}}\right] \binom{x_{n}}{u_{n}} + \left(\frac{\omega_{n,x}^{2}}{\varepsilon_{x}^{2}}\right) x_{n,t} + \left(\frac{2\zeta_{x}\omega_{n,x}}{\varepsilon_{x}}\right) u_{n,t} + a_{x,n,r}$$

$$a_{y,n} = -\left[\frac{\omega_{n,y}^{2}}{\varepsilon^{2}} \quad \frac{2\zeta_{y}\omega_{n,y}}{\varepsilon}\right] \binom{y_{n}}{v_{n}} + \left(\frac{\omega_{n,y}^{2}}{\varepsilon^{2}}\right) v_{n,r} + \left(\frac{2\zeta_{y}\omega_{n,y}}{\varepsilon}\right) v_{n,r} + a_{y,n,r}$$

$$a_{z,n} = -\left[\frac{\omega_{n,z}^{2}}{\varepsilon^{2}} \quad \frac{2\zeta_{z}\omega_{n,z}}{\varepsilon}\right] \binom{z_{n}}{w_{n}} + \left(\frac{\omega_{n,z}^{2}}{\varepsilon^{2}}\right) z_{n,r} + \left(\frac{2\zeta_{z}\omega_{n,z}}{\varepsilon}\right) w_{n,r} + a_{z,n,r}$$

$$\varepsilon_{x} = \varepsilon_{y} = \varepsilon_{z} = 1$$

$$\zeta_{x} = 1, \quad \zeta_{y} = 1, \quad \zeta_{z} = 1.1$$

$$\omega_{n,x} = 0.54, \quad \omega_{n,y} = 0.62, \quad \omega_{n,z} = 0.78$$

Outer-loop control system gain and phase margins

Outer-loop control system gain and phase margins

Outer-loop control system gain and phase margins

UAV Control System Design



Control performance



Single-rotor Helicopter Coaxial Helicopter

Quadrotor

The END!



